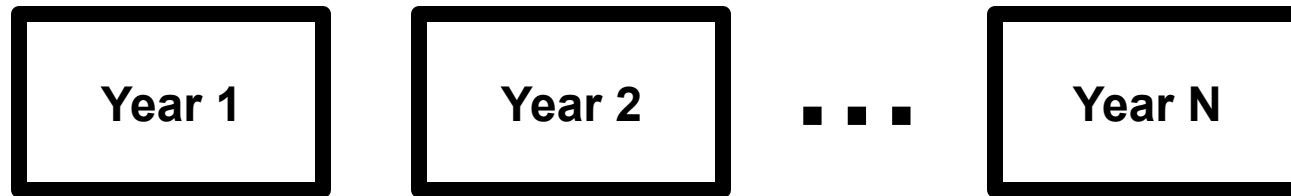


Multi-period planning models under uncertainty



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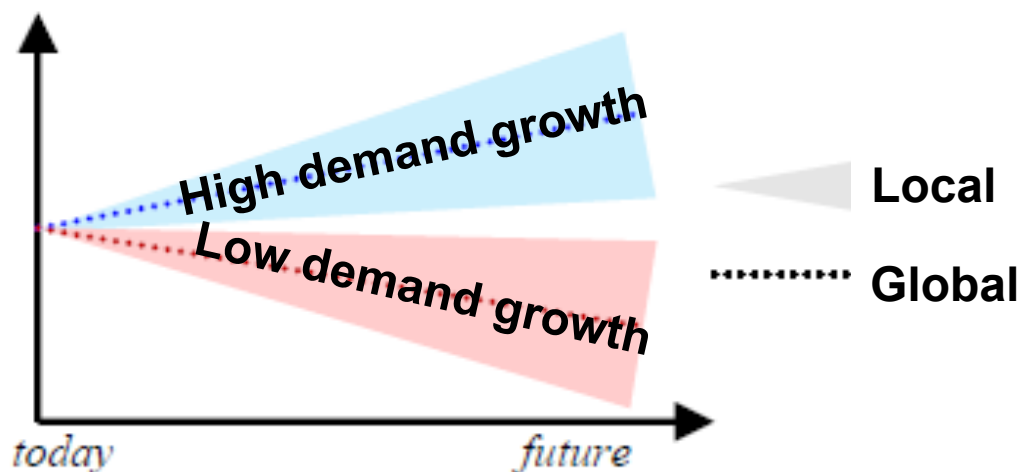
March 20 , 2013
GRID TRANSFORMATION WORKSHOP
Portland Oregon

Overview

1. Types of uncertainty
 - a. Local uncertainty
 - b. Global uncertainty
2. Types of models
 - a. Assessment models
 - b. Design models
3. An assessment model: guided Monte Carlo
4. Design models
 - a. For local uncertainties
 - b. For global uncertainties

Types of uncertainty

- a. Global - uncertainties for which different values produce dramatically different results: emissions policies, large demand shifts, coal or nuclear unavailability, extremes in fuel prices, extended drought, dramatic change in technology investment costs.
- b. Local – uncertainty in values a parameter may take under a global realization: variation in load growth, investment costs or fuel prices.



Category B & C contingencies can be treated as certainties, i.e., we plan as if they will definitely occur, or they can be treated as local uncertainties, with assigned probabilities. .

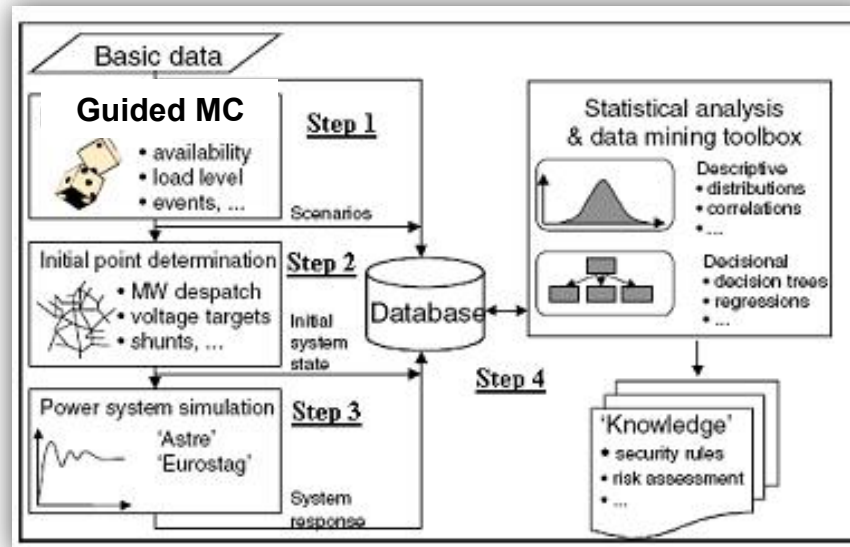
Types of models

- a. **Assessment – these provide performance measures/indices: Power flow, time-domain simulation, production cost, reliability evaluation (Monte Carlo, enumeration).**
- b. **Decision/Design – these provide investment solutions that are “best” under the range of uncertainties: generation expansion planning (GEP), transmission expansion planning (TEP).**
 - **Decision/Design models are used for GEP (e.g., EGEAS) but very little for TEP.**
 - **GEP models under uncertainty are research-grade.**
 - **Deterministic TEP models are research-grade.**
 - **TEP models under uncertainty under development.**

An assessment model: Guided MC

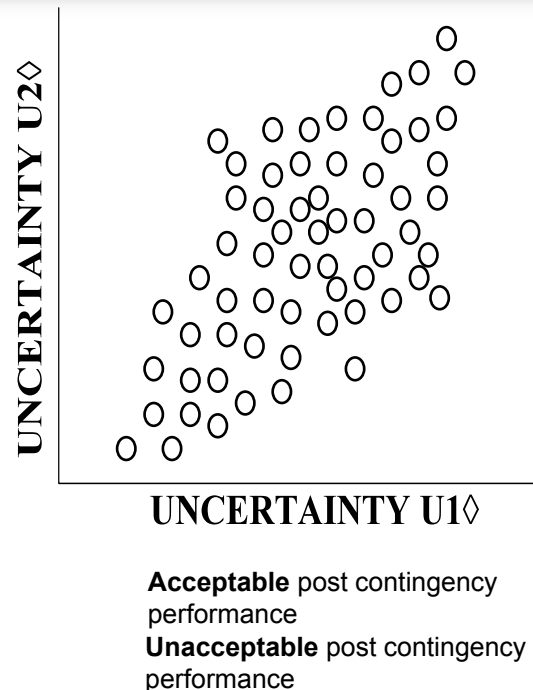
DATABASE DEVELOPMENT

1. Guided (importance) sampling
2. Optimal power flow
3. Contingency analysis



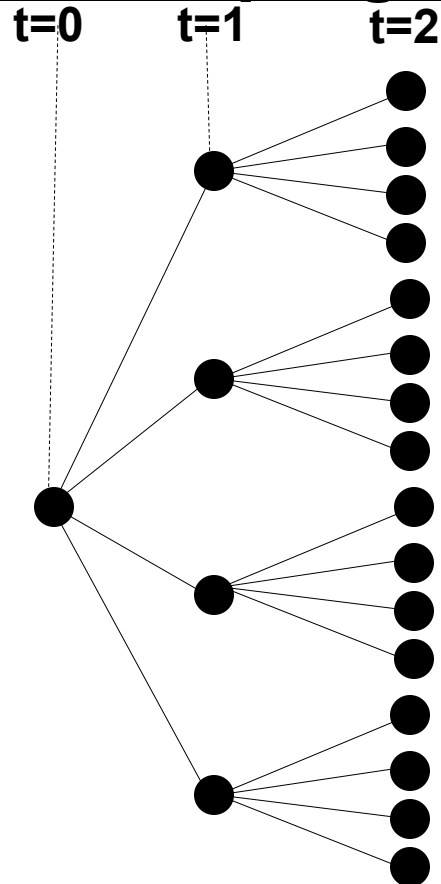
4. STATISTICAL ANALYSIS

- ☐ Estimate reliability indices
 - ☐ $prob(load\ loss)$
 - ☐ $prob(unacceptable\ perf)$
 - ☐ Risk
 - ☐ ...
- ☐ Analyze Relationships
 - ☐ Distributions
 - ☐ Correlations



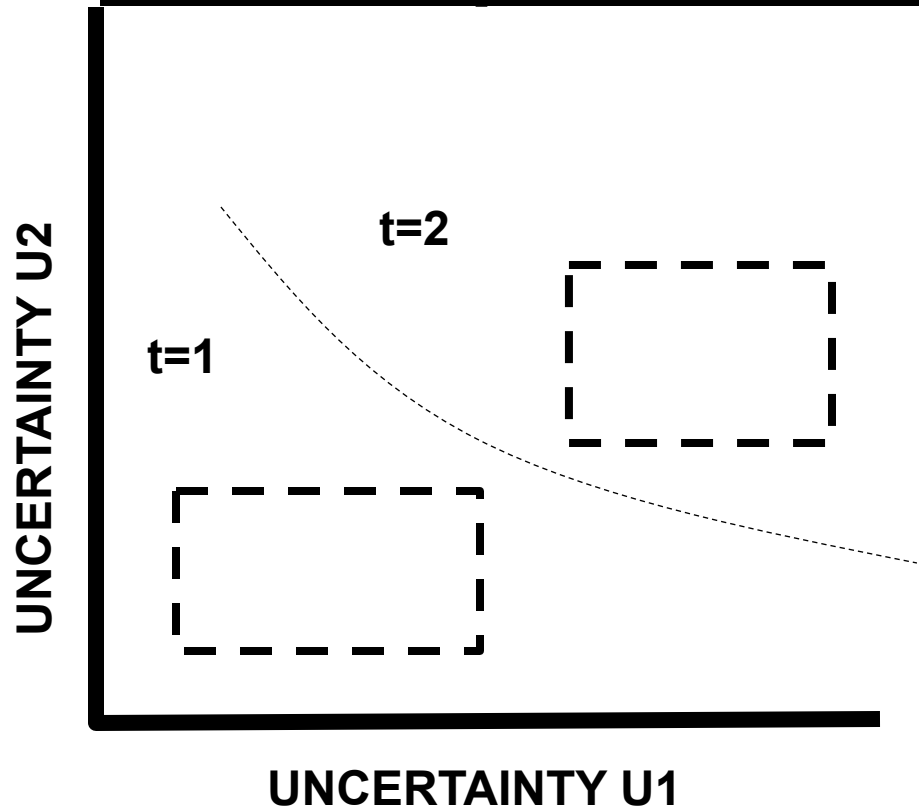
Design models for local uncertainties

Stochastic programming



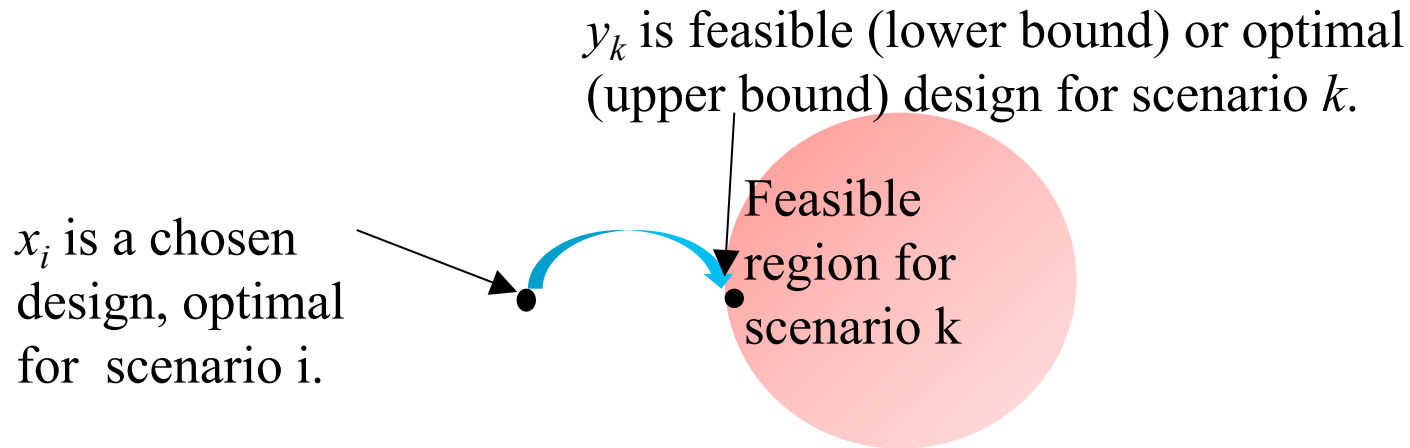
- **Weaknesses:** Cannot have too many uncertainties; need prob distributions.
- **Strengths:** Recourse counters conservatism; SP may also be applicable for global uncertainties.

Robust optimization



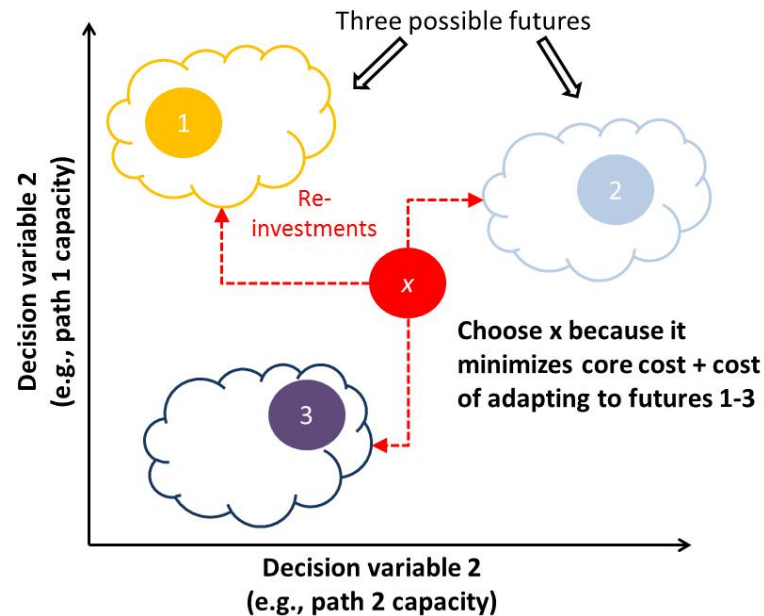
- **Weaknesses:** Results in solution that is feasible for entire uncertainty range and therefore conservative.
- **Strengths:** May be computationally more tractable. Need bounds, not prob distributions.

A design model for global uncertainty



The adaptation cost of x_i to scenario k is the minimum cost to move x_i to a feasible or optimal design y_k in scenario k . It measures the cost of our plan x_i if scenario k happens.

A design model for global uncertainty



Identifies an investment that is “core” in that the total “CoreCost” plus the cost of adapting it to the set of envisioned futures is minimum.

Minimize:

$$\text{CoreCosts}(\underline{x}^f) + \beta [\sum_i \text{AdaptationCost}(\Delta \underline{x}_i)]$$

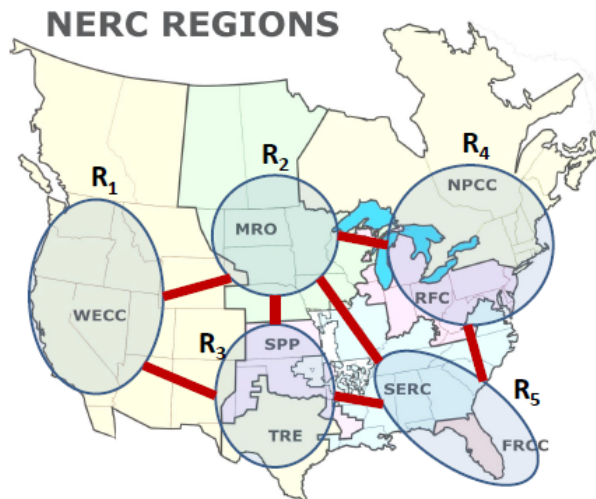
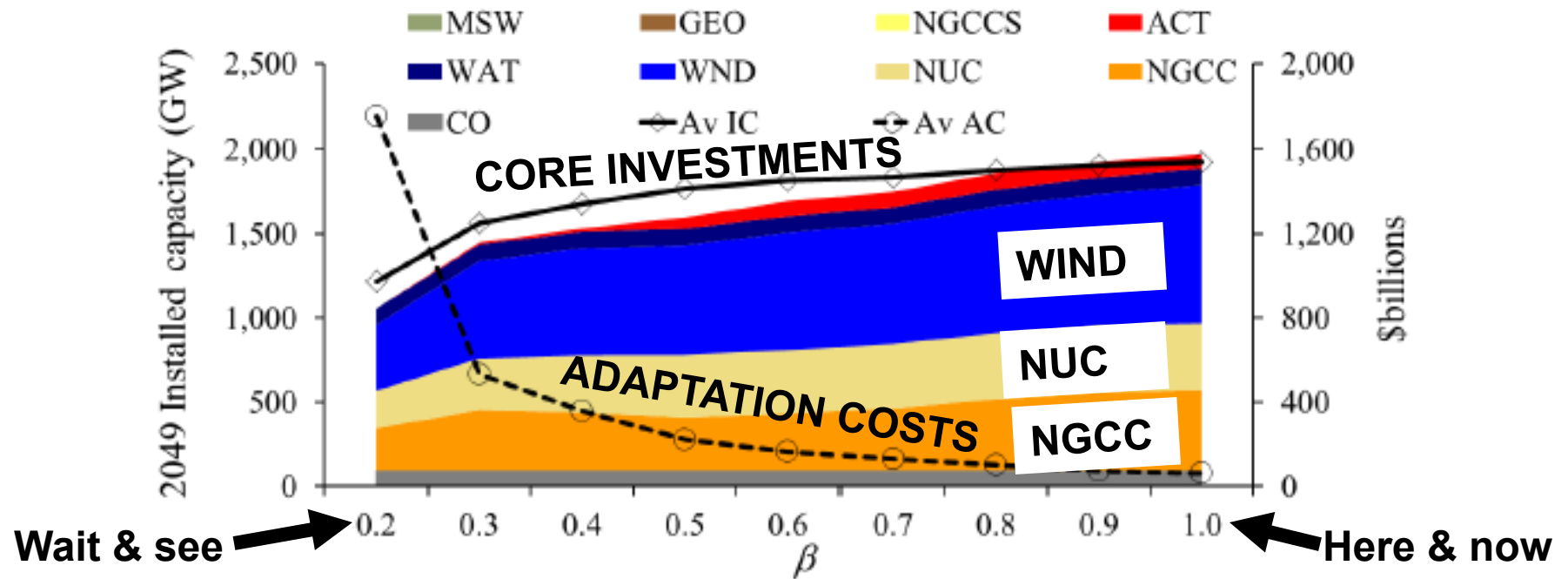
Subject to:

$$\text{Constraints for scenario } i=1, \dots, N: \underline{g}_i(\underline{x}^f + \Delta \underline{x}_i) \leq \underline{b}_i$$

\underline{x}^f : Core investments, to be used by all scenarios i

$\Delta \underline{x}_i$: Additional investments needed to adapt to scenario i

A design model for global uncertainty



Minimize:

$$\text{CoreCosts}(\underline{x}^f) + \beta \left[\sum_i \text{AdaptationCost}(\Delta \underline{x}_i) \right]$$

Subject to:

$$\text{Constraints for scenario } i=1, \dots, N: g_i(\underline{x}^f + \Delta \underline{x}_i) \leq \underline{b}_i$$

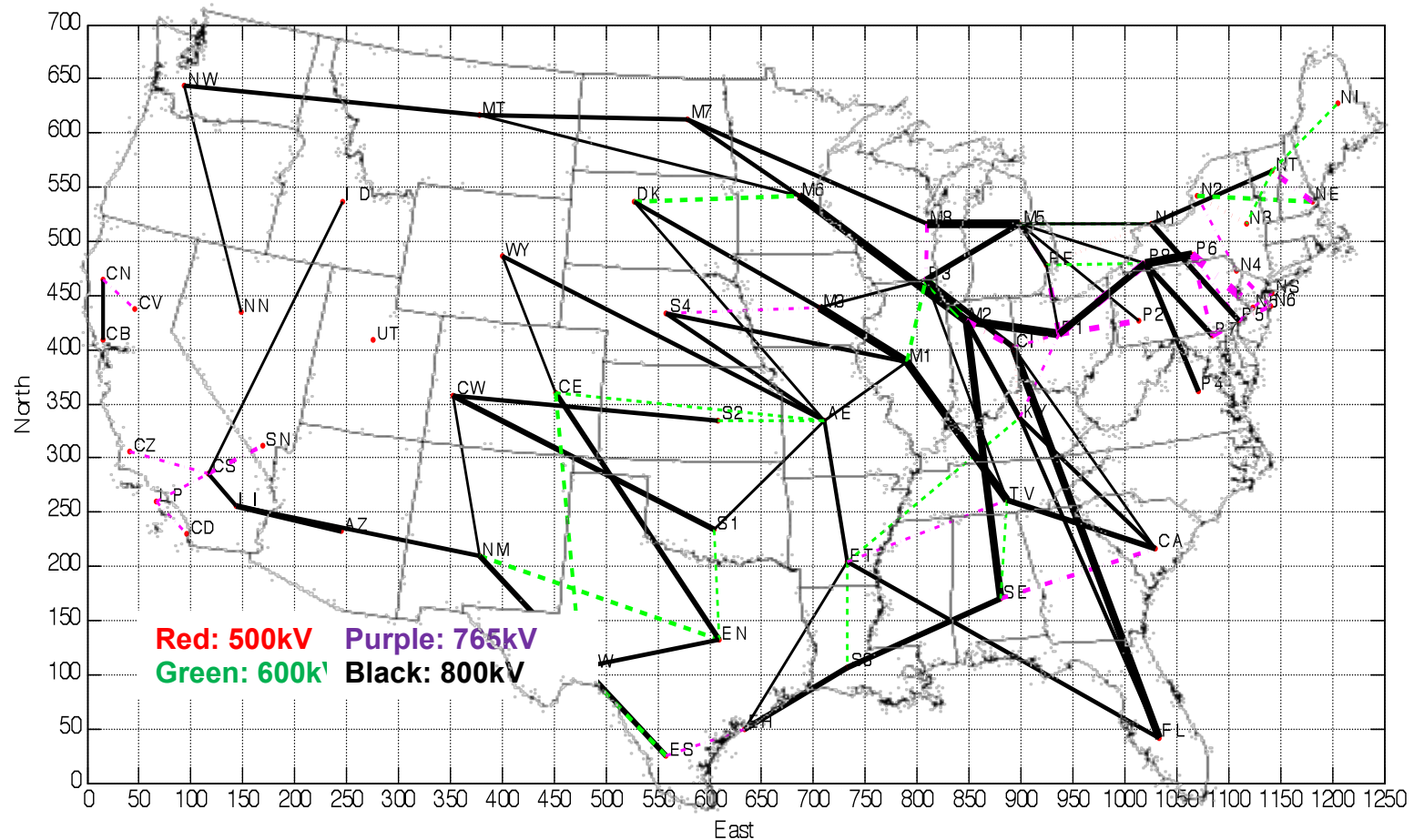
\underline{x}^f : Core investments, to be used by all scenarios i

$\Delta \underline{x}_i$: Additional investments needed to adapt to scenario i

Final comments

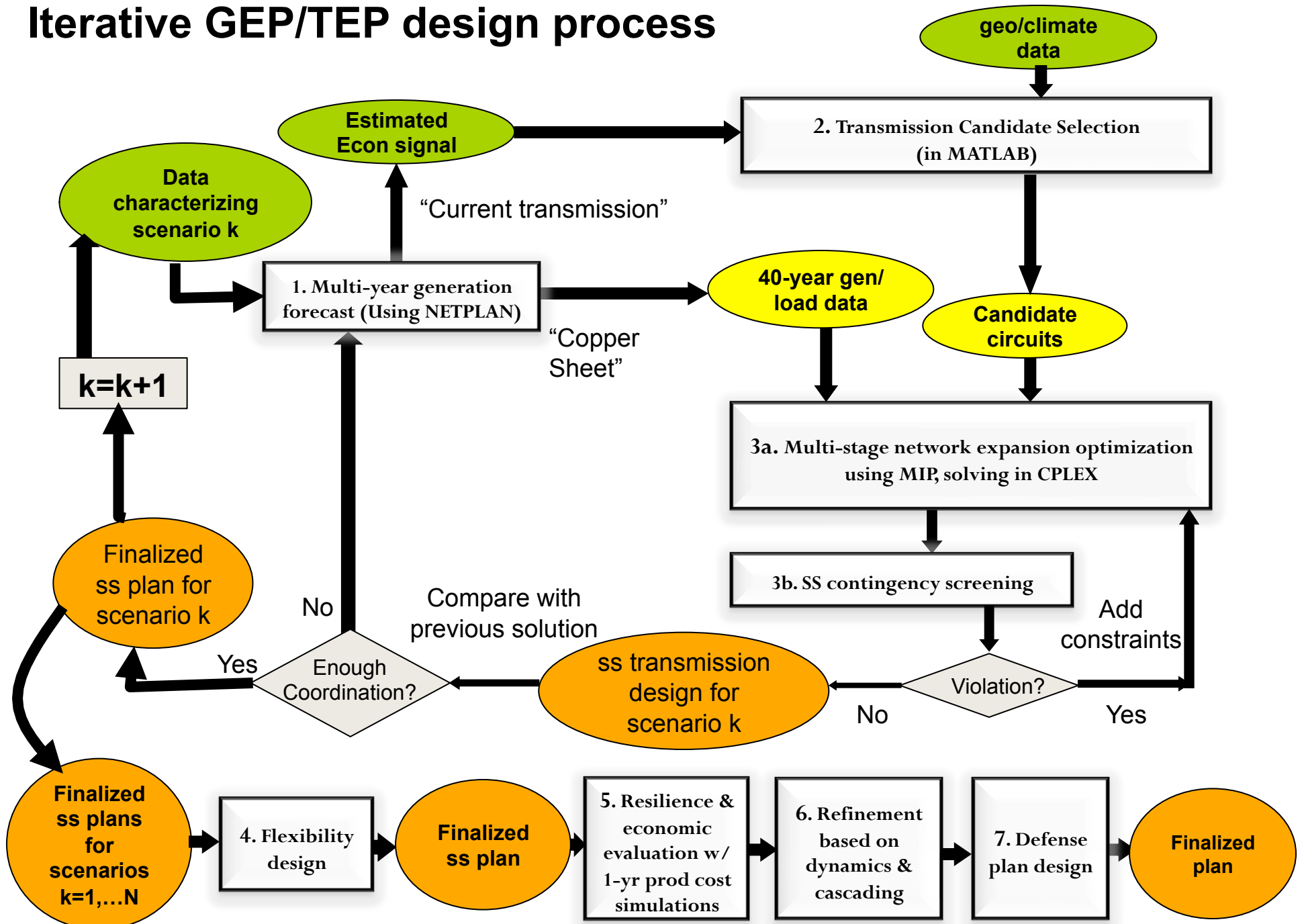
1. Tools for transmission assessment under uncertainty are available and are used.
2. Guided Monte-Carlo: a proven & effective way to harness computation for design exploration
3. Tools for deterministic transmission design are just now becoming available, but transmission design under uncertainty is a research area.
4. Local uncertainties: addressed by SP and RO.
5. Global uncertainties: Design for flexibility!
6. SP, RO, and Flexibility design needs resources to be developed into commercial tools.

A deterministic TEP result



A renewable-heavy generation portfolio, with particular emphasis on wind in the Midwest and Northwest and geothermal in the West.

Iterative GEP/TEP design process



Scenarios used in flexibility design

Cluster	Gas price	Gas production limits	Demand	RPS	CO ₂ ^{cap}	Wind investment cost
	GP	GPL				WC
Benchmark	L	No	L	No	No	H
1	L	No	L	No	No	L
2	L	No	L	Yes	No	H
3	L	No	L	Yes	Yes	L
4	L	No	H	No	Yes	H
5	L	Yes	L	Yes	Yes	H
6	H	No	L	Yes	Yes	L
7	H	No	H	No	Yes	L
8	H	Yes	L	No	No	L
9	H	Yes	L	Yes	Yes	H
10	H	Yes	H	Yes	No	H